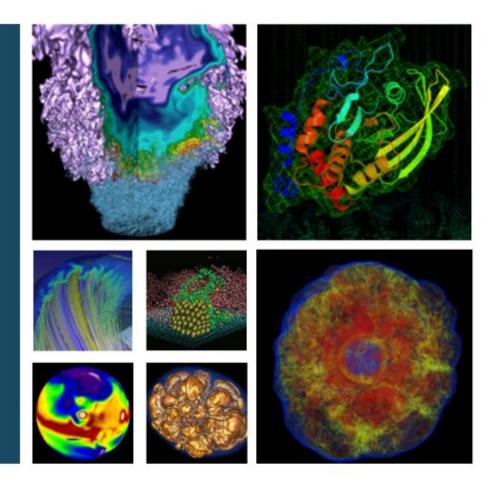
Using Spark on Cori





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What is Spark?



- Apache Spark is a fast and general engine for large-scale data processing
- Data-parallel model
 - Similar to Hadoop, except it harnesses in-memory data for fast data processing
- Robust body of libraries: machine learning, sql, graphx, streaming
- Created at the AMPLab in UCB, currently maintained by DataBricks, more than 400 contributors to git repository
- Requires no parallel programming expertise: Can take Python, Java, or Scala code and automatically parallelize it to an arbitrary number of nodes

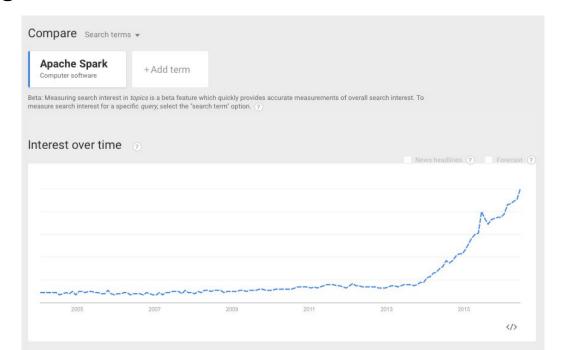




Spark: So hot right now



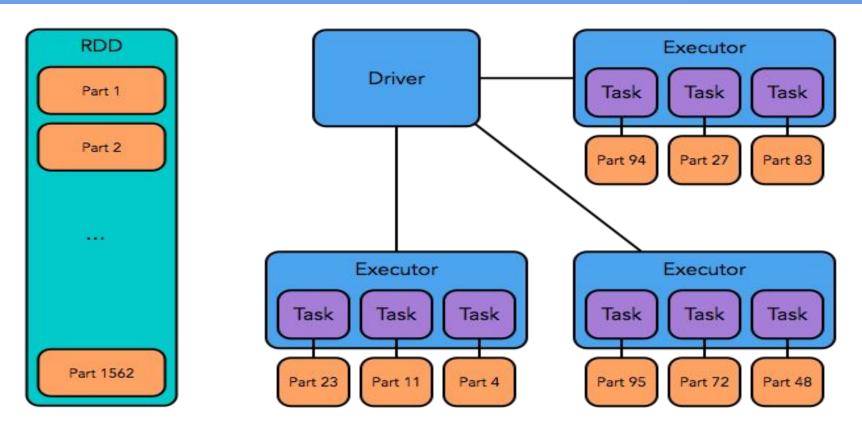
- Spark Summit in June had more than 2500 attendees
- Used by Airbnb, amazon, eBay, Groupon, IBM,
 MyFitnessPal, NetFlix, OpenTable, Salesforce, Uber, etc.
- And by NASA (SETI), UCB (ADAM)
- Used at NERSC for metagenome analysis, astronomy, and business intelligence





Spark Architecture





- One driver process, many executors
- Driver creates DAG and coordinates data and communication (via akka TCP)
- Datasets generally cached in memory, but can spill to local disk



Spark API



- Can use Java, Python, or Scala
- Latest version also has an R interface and a Jupyter notebook
- Most newer functionality comes to Scala interface first
- Also some indications that Scala performs better
- Python is easier to use, so we recommend using that to start out
 - Can always reoptimize in scala later if necessary
- Default version at NERSC is Spark 2.0.0
- Recommend running on Cori
 - Larger memory nodes and faster connections with scratch file system

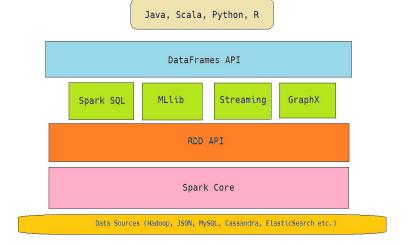


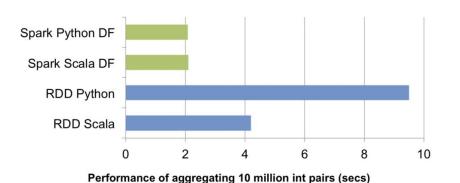


Spark Data Formats



- RDD: Resilient Distributed Dataset. Immutable collection of unstructured data.
- DataFrames: same as RDD but for structured data.
 Familiar pandas-like interface
- Which to use?
 - Future: DataFrames
 - Now: DataFrames and fall back to RDDs for more fine tuned control









Spark Goodies



- Streaming: high-throughput, fault-tolerant stream processing of live data. Twitter, kafka, etc.
- SQL: Structured data processing, lives on top of RDD API
- ML: classification, regression, clustering, collaborative filtering, feature extraction, transformation, dimensionality reduction, and selection. Best only for really LARGE datasets. Demo tomorrow
- GraphX: Graph parallel computation. Extends RDD to a directed multigraph with properties attached to each vertex and edge





Spark Strategy



- Best for DATA PARALLEL!!!
- Make sure you have enough memory
 - Datasets are held in memory for speed
 - When memory is filled up, must spill data to disk, which slows things considerably
 - NERSC default spills to RAM file system and to Lustre scratch
 - If you get "java.lang.OutOfMemoryError" your executors have run out of memory. Fix by making the load on each node smaller. Either
 - Increase the number of nodes OR
 - Increase executor memory

Keep the cores fed

 Optimal number of partitions is 2 - 3 times the number of cores, automatic partitioning sizes can be a little wonky



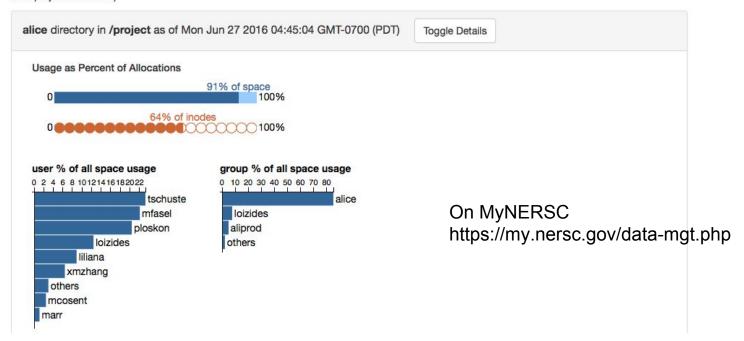


Real World Use Case: Data Dashboard



Data Dashboard

Showing disk space and inode usage for project directories at NERSC to which you have access as PI, PI proxy, or user (includes /project, /projecta, and /projectb/sandbox)







Conclusion



- Spark is an excellent tool for data parallel tasks that are too big to fit on a single node
 - Easy to use python DataFrames interface
 - Does parallel heavy lifting for you
- Spark 2.0 installed at NERSC, performs well on data-friendly Cori system
- Come to the hack session tomorrow for more Spark demos!







National Energy Research Scientific Computing Center





Spark Batch Job



#!/bin/bash

```
#SBATCH -p debug

#SBATCH -N 2

#SBATCH -t 00:30:00

#SBATCH -e mysparkjob_%j.err

#SBATCH -o mysparkjob_%j.out

#SBATCH --ccm
```

module load spark

start-all.sh spark-submit --master \$SPARKURL --driver-memory 15G --executor-memory 20G ./testspark.py stop-all.sh

Submit with sbatch <submitscriptname.slurm>





Spark Terminology



- Driver process: runs on head node and coordinates workers (you can adjust memory allocation)
- Executor: single process that runs on worker nodes
- Task: a unit of work that's sent to one executor
- Application: entire program that is run
- Job: spawned by a spark action (like collect), consists of one or more stages on multiple nodes
- Stage: a set of tasks that do the same operation on a different slice of the data
 - See frequent references to "job" and "stage" in the spark logs
- RDD (Resilient Distributed Datasets): read only collection of records, held in memory or spilled out to disk



